

# Understanding Haddock somatic growth changes on Eastern Georges Bank

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## Part 1. Modelling haddock growth changes

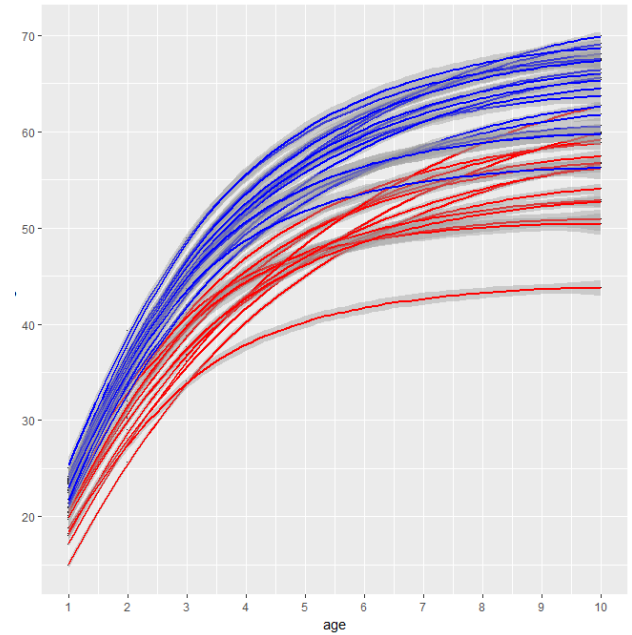
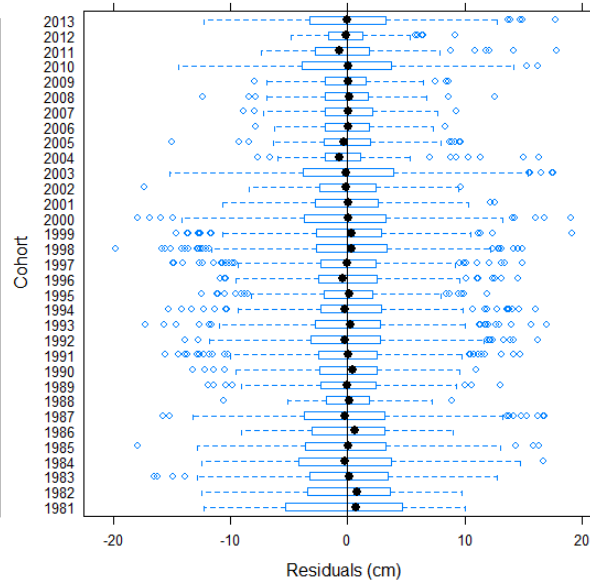
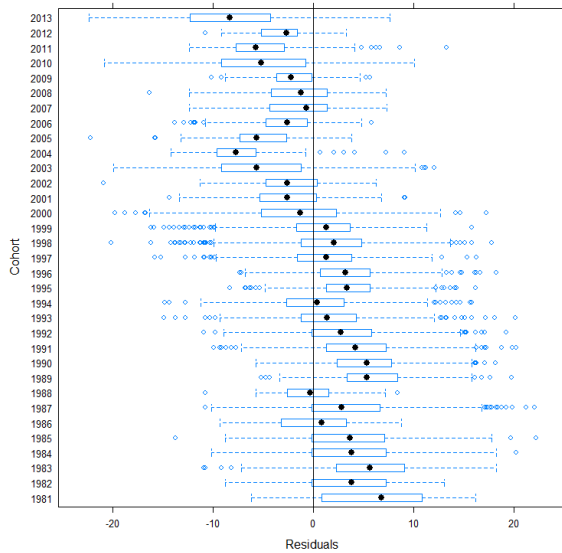
## Haddock length data

- EGB haddock length at age from 1986-2017 using DFO survey samples
- Ages 1-8

# Von Bertalanffy Mixed effect model

$$L_a \sim \text{Linf} * (1 - \exp(-1 * K * (\text{Age} - t_0)))$$

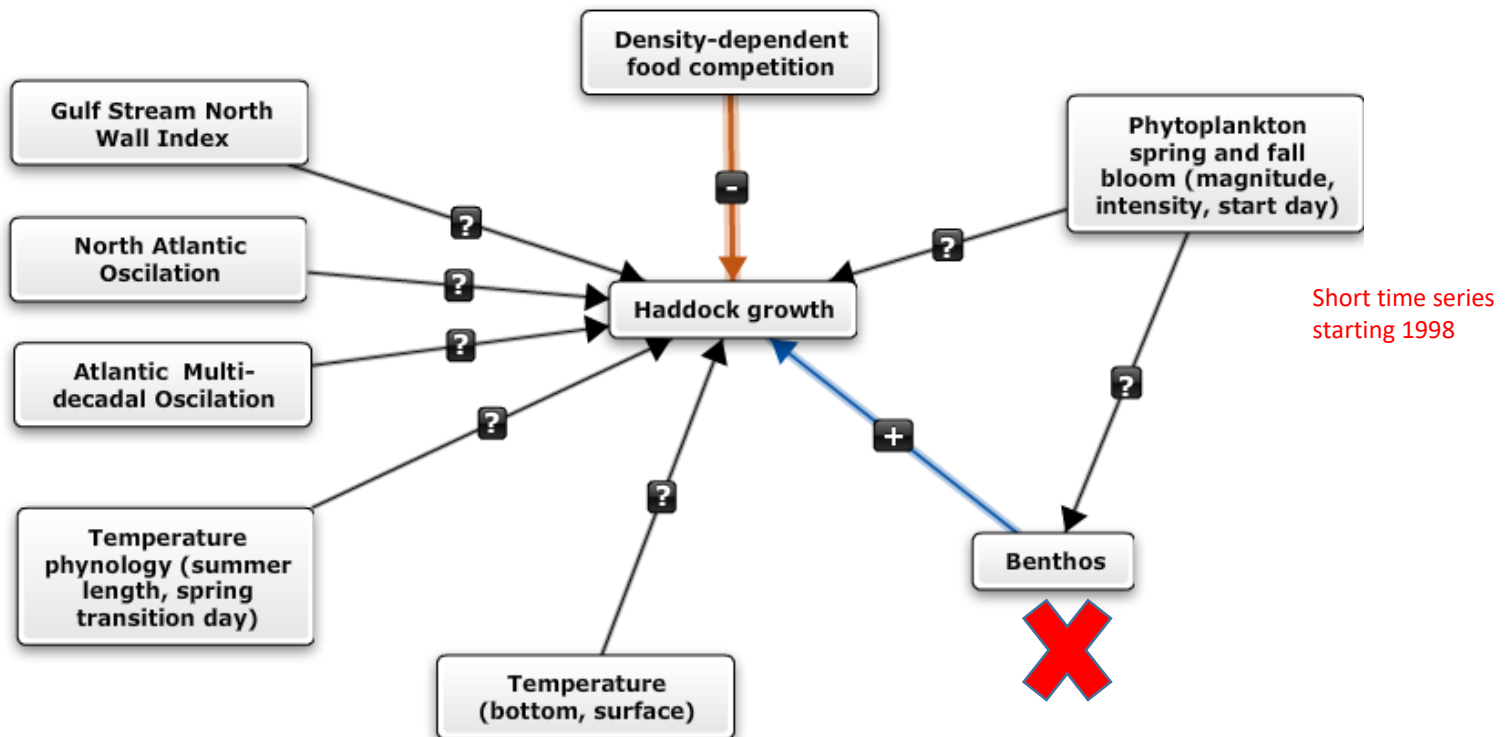
## Von Bertalanffy Model



## Part 2.

Understanding the mechanism of somatic growth  
changes of Eastern Georges Bank haddock

# Conceptual model: possible relevant factors to haddock growth



# Fish Density

## ➤ *Cohort strength:*

$$\ln A_{ijs} = \beta_i + \beta_j + \beta_s + \beta_{is} + \epsilon$$

where  $A_{ijk}$  is the survey index at age  $i$ ,  $i = 1, 2$ , and cohort  $j$ ,  $j = 1985, 1986, \dots, 2015$ , in survey  $s$

## ➤ *Annual biomass:*

$$\ln B_{ts} = \beta_t + \beta_s + \epsilon$$

where  $B_{ts}$  is the survey biomass index in year  $t$ ,  $t = 1987, \dots, 2017$  from survey  $s$

# GAM model

For both long(1987-2018) and short(1998-2018) time series data:

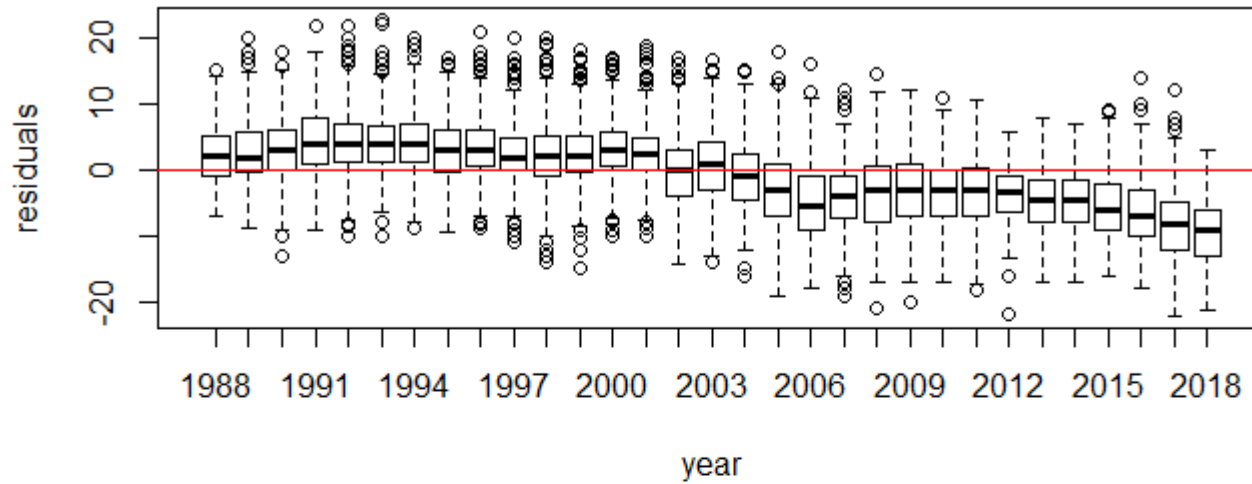
- *Step 1:* GAM model to relate Length with Age

$$La \sim s(\text{age}) + \epsilon \quad \epsilon \sim N(0,1)$$

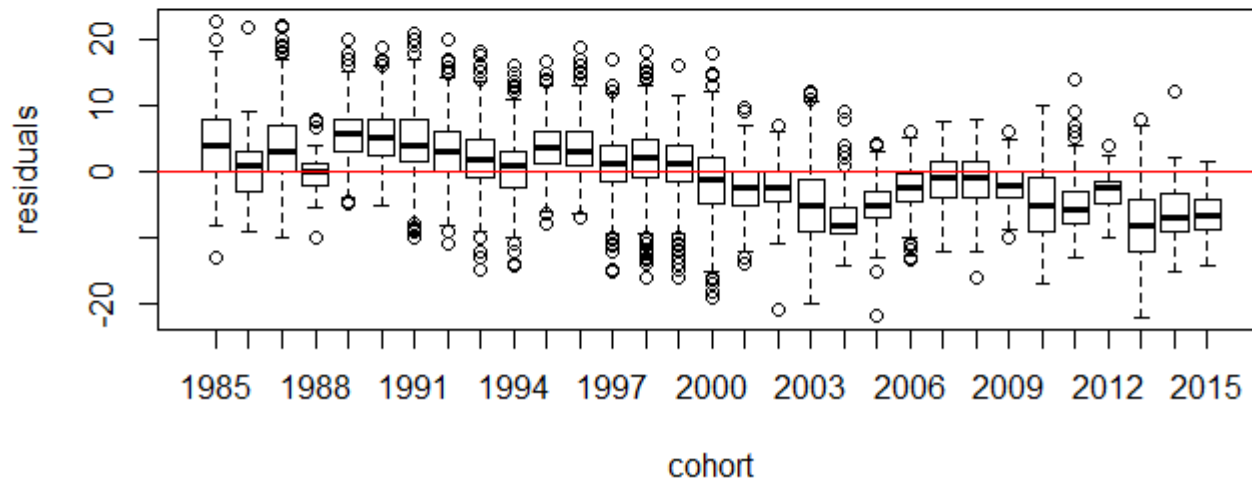
- *Step 2:* GAM model to interpret the impact of environmental factors on the variation of haddock growth.
  - Response variable: residuals ( $\epsilon$ ) from step 1.
  - Predictor variables: variables in the conceptual model except for benthos data



## Long time series data

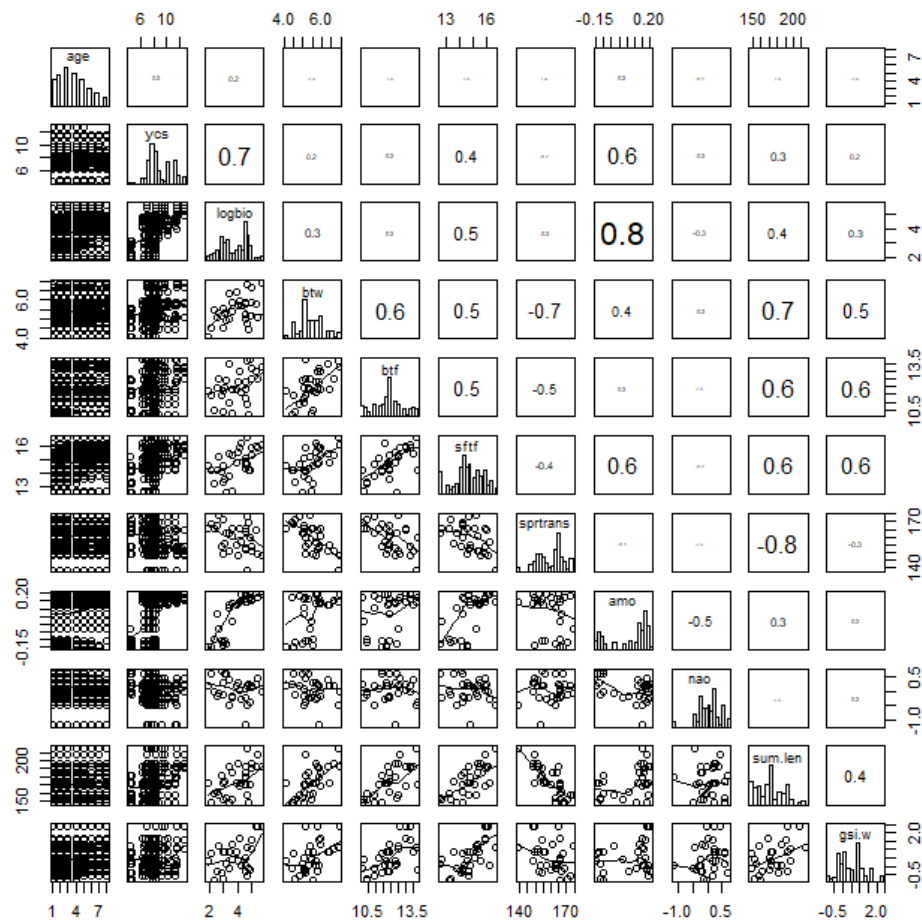


$$La \sim s(\text{age}) + \epsilon$$



# Collinearity check

data\_sh3,all in GBK



- Covariate variables with the variance inflation factor ( $VIF > 3$ ) and linear correlation  $r^2 > 0.7$

# GAM models

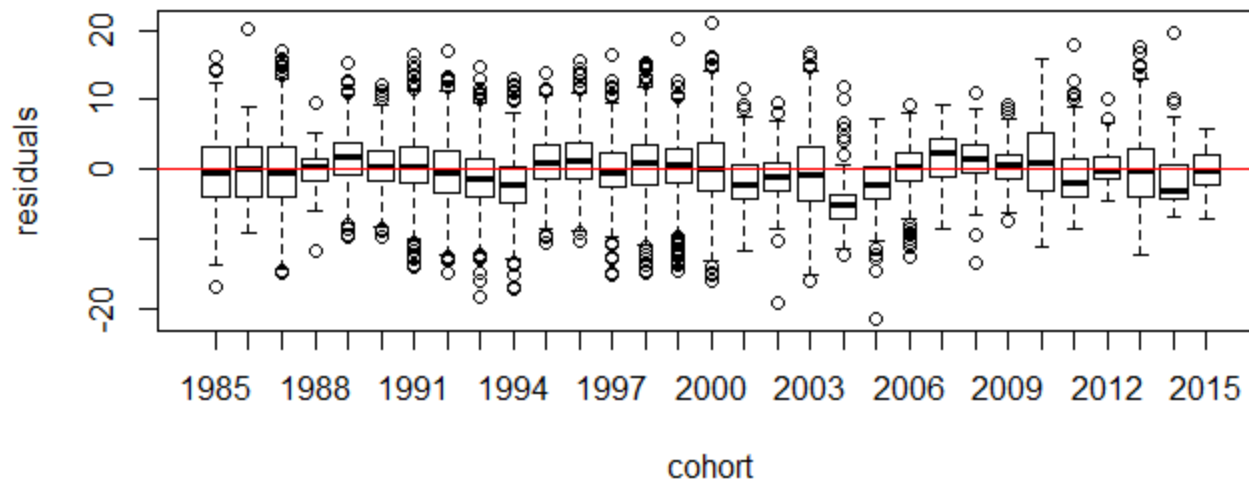
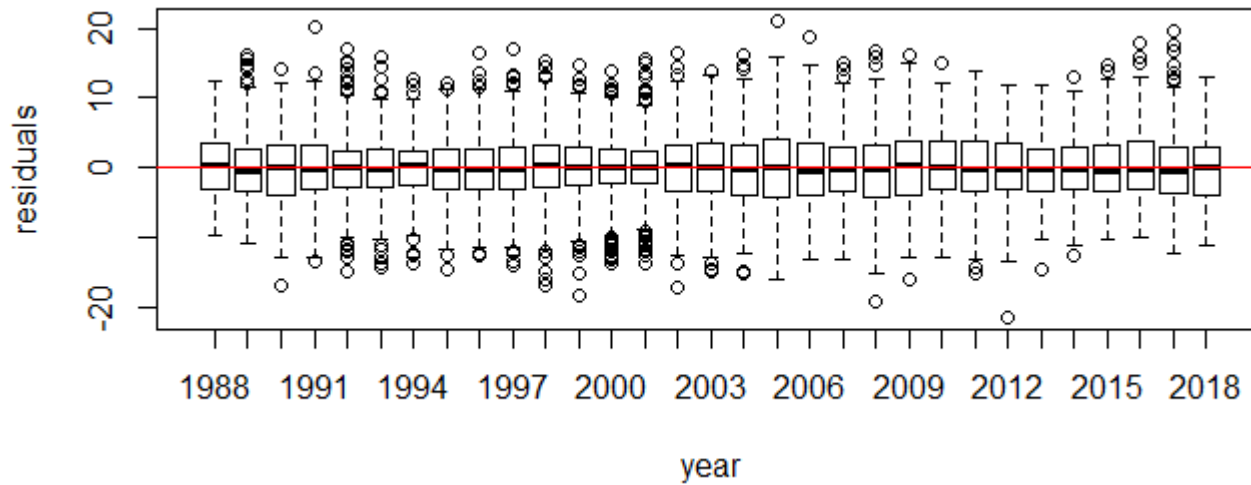
➤ **Smooth parameter estimation and variable selection:** Double penalty(Marra and Wood,2011 ) smooth approach.

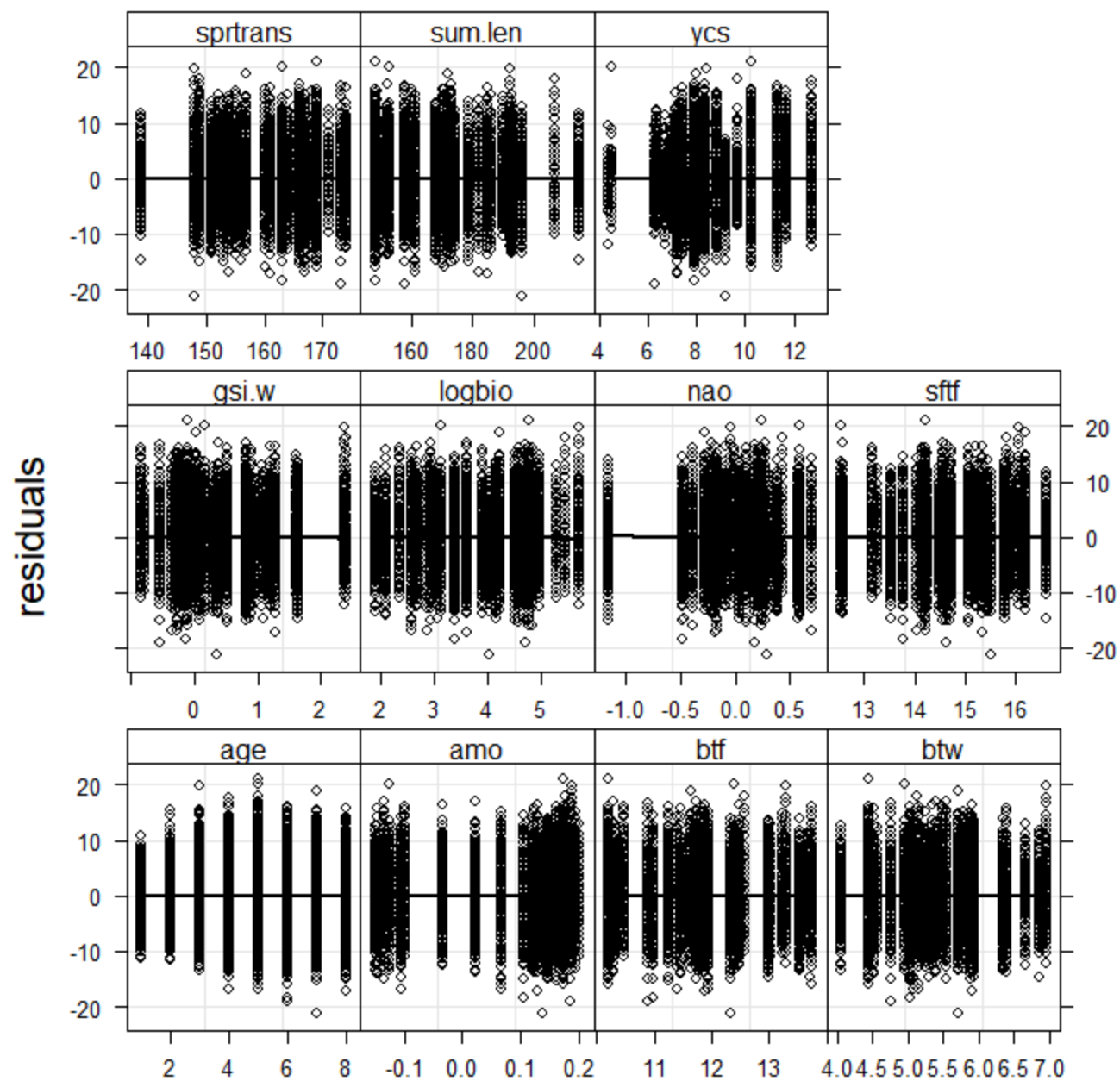
➤ ***Model selection: based on AIC and anova test***

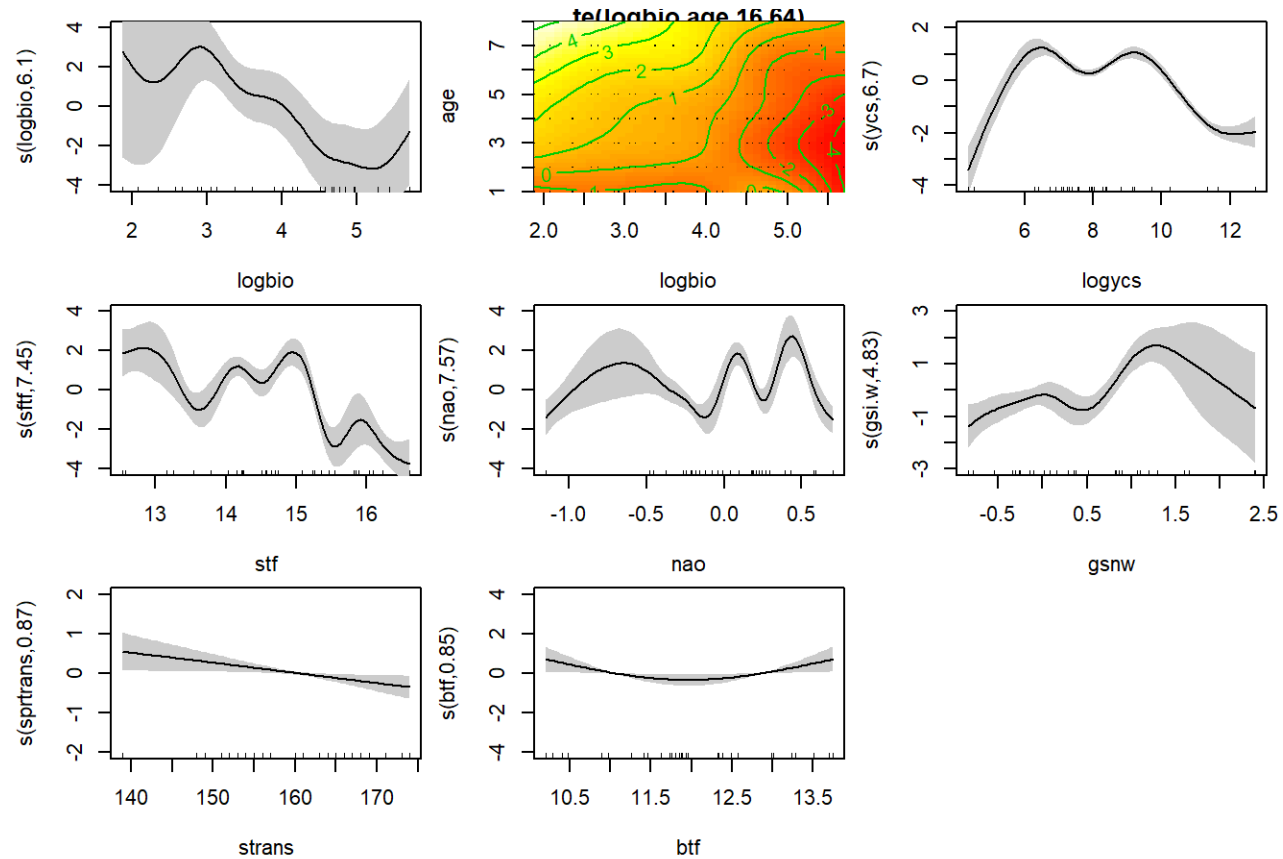
- $\epsilon \sim s(\text{ycs})+s(\text{bio})+s(\text{btf})+s(\text{sftf})+s(\text{sprtrans})+s(\text{nao})+s(\text{gsi.w})$
- $\epsilon \sim s(\text{ycs})+ s(\text{bio})+ \text{te}(\text{bio, age})+s(\text{btf})+s(\text{sftf})+s(\text{sprtrans})+s(\text{nao})+s(\text{gsi.w})$
- $\epsilon \sim s(\text{ycs})+s(\text{bio, by=age})+s(\text{btf})+s(\text{sftf})+s(\text{sprtrans})+s(\text{nao})+s(\text{gsi.w})$

# Model check

$$\epsilon \sim s(\text{yco}) + s(\text{bio}) + \text{te}(\text{bio}, \text{age}) + s(\text{btf}) + s(\text{sftf}) + s(\text{sprtrans}) + s(\text{nao}) + s(\text{gsi.w})$$

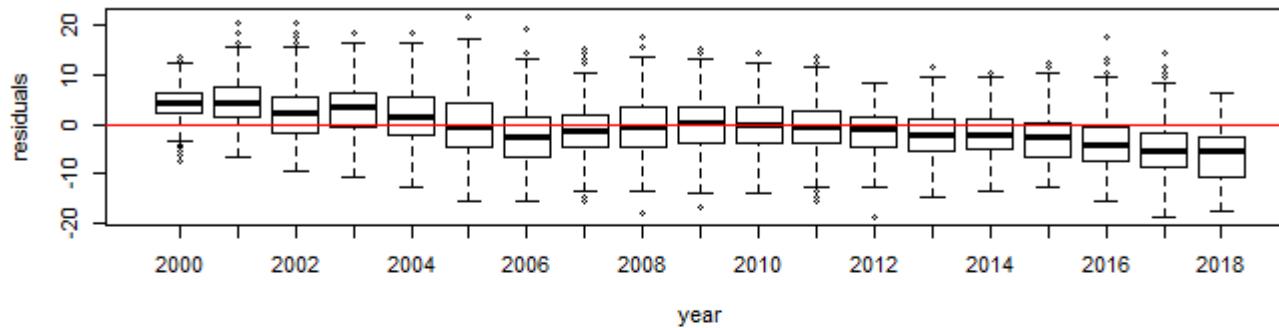




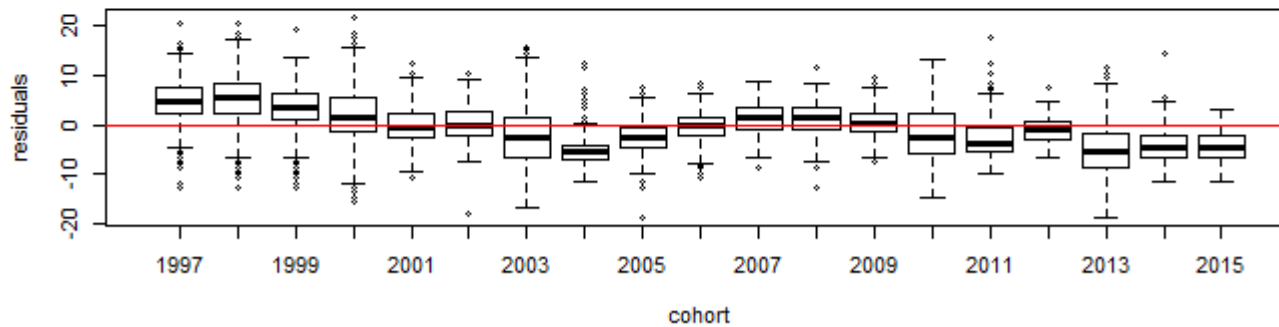


- Deviance explained 40.6%
- density-dependent effect explained 36.8% of deviance

# short time series data(1998-2018)



$$La \sim s(\text{age}) + \epsilon$$



# GAM models

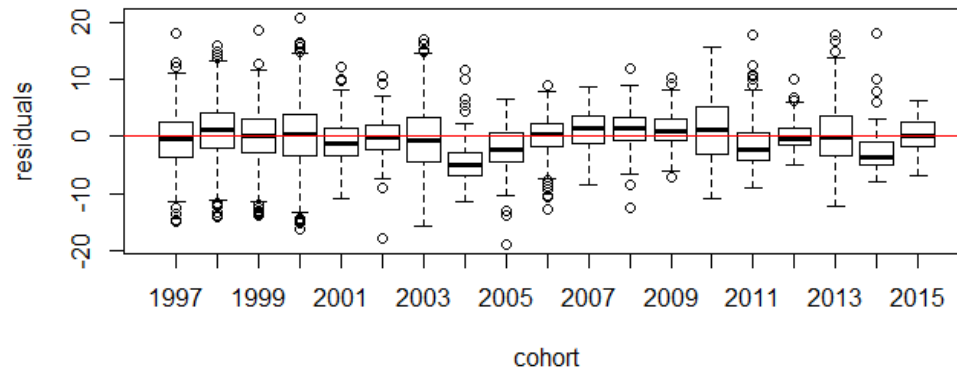
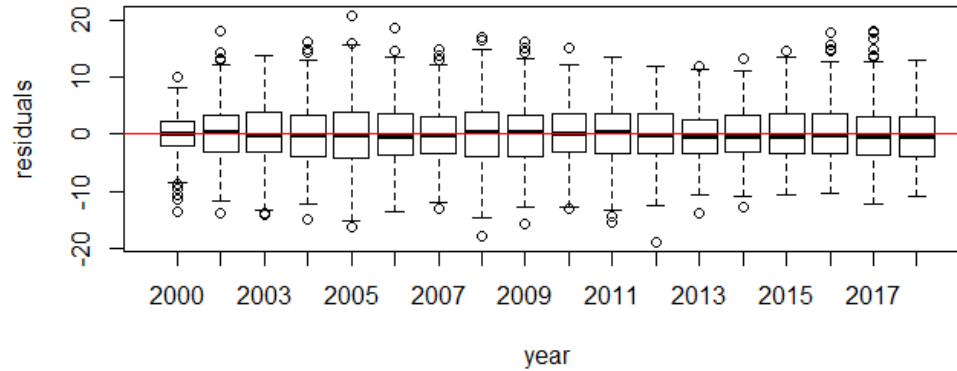
➤ **Model selection:** based on AIC and anova test

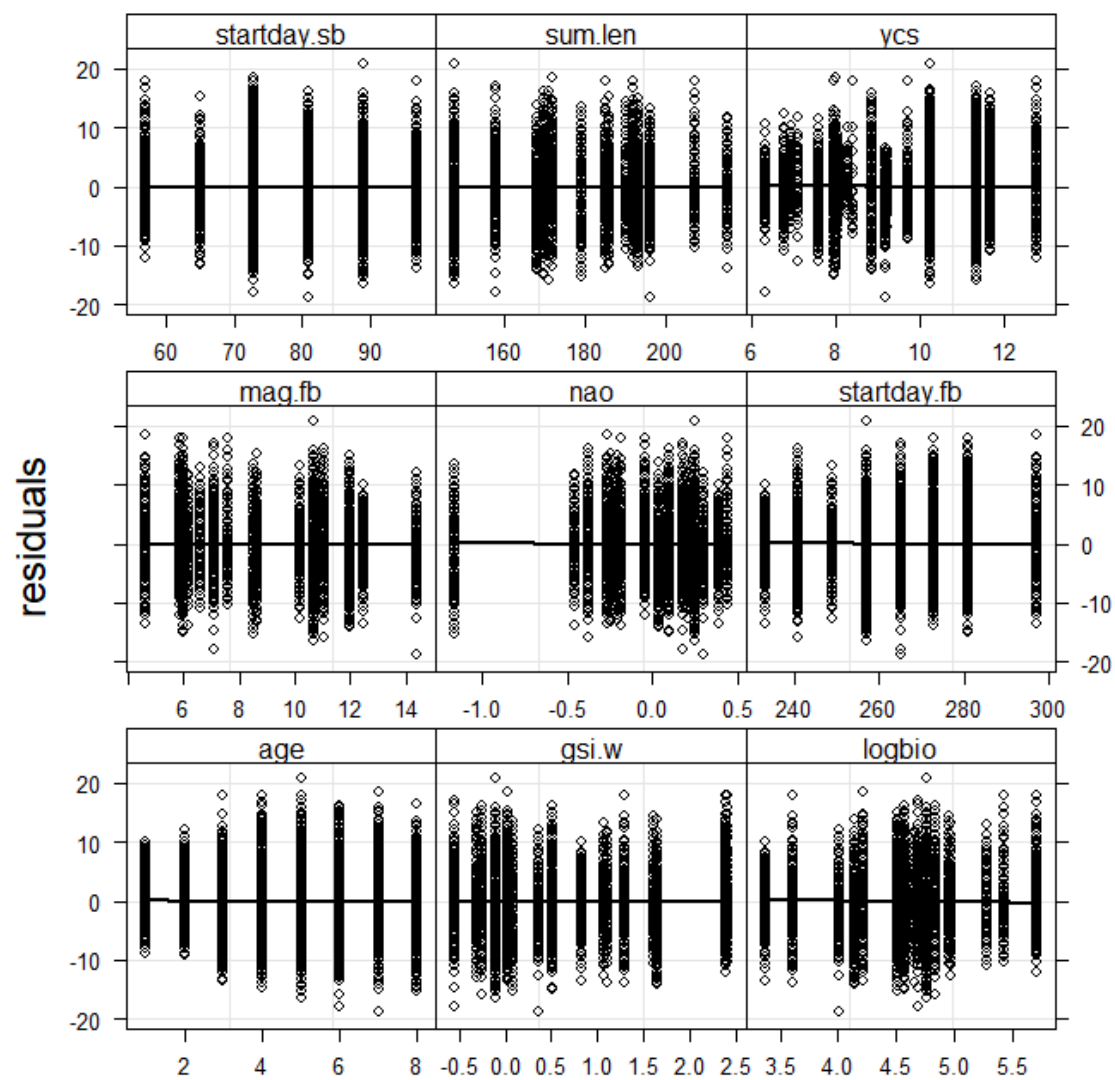
- $\epsilon \sim s(\text{y cs}) + s(\text{bio}) + s(\text{btf}) + s(\text{sftf}) + s(\text{sprtrans}) + s(\text{nao}) + s(\text{gsi.w})$
- $\epsilon \sim s(\text{y cs}) + s(\text{bio}) + \text{te}(\text{bio}, \text{age}) + s(\text{btf}) + s(\text{sftf}) + s(\text{sprtrans}) + s(\text{nao}) + s(\text{gsi.w})$
- $\epsilon \sim s(\text{y cs}) + s(\text{bio}, \text{by}=\text{age}) + s(\text{btf}) + s(\text{sftf}) + s(\text{sprtrans}) + s(\text{nao}) + s(\text{gsi.w})$

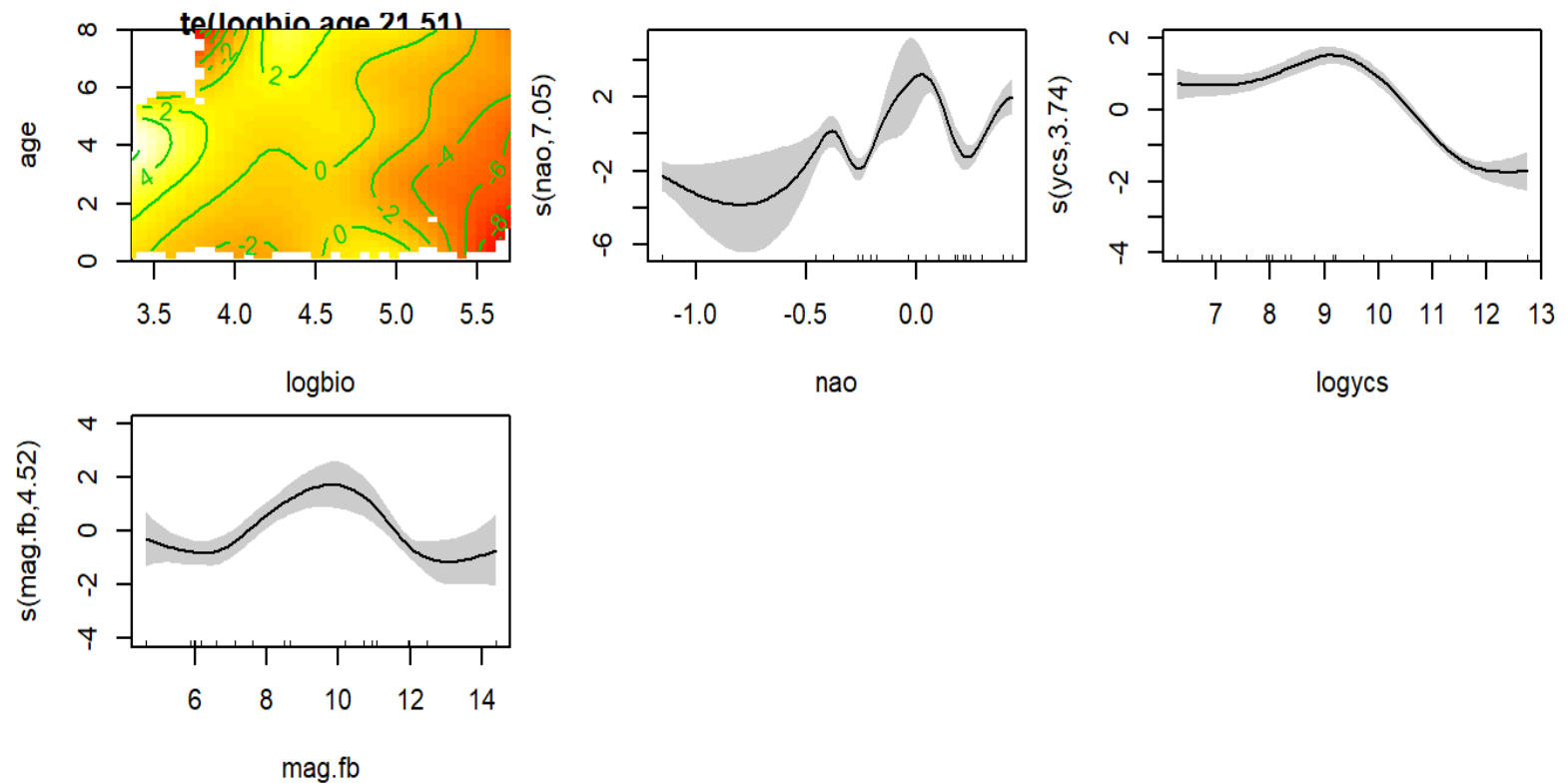


# Model check

$$\epsilon \sim s(\text{yco}) + \text{te}(\text{bio}, \text{age}) + s(\text{nao}) + s(\text{mag.fb})$$







- Deviance explained 23.6%
- density-dependent effect explained 19.1% of deviance

# summary

- Density dependent effect and possible high temp in the summer and fall have the most influences on haddock growth changes.
- Consistent with Clarks(1969) description about fish size reduction following very strong cohorts in the history.
- Care is needed with the interpretation of the very weak cohorts at older ages due to small number of samples
- When there are high concavity among covariate variables, accurately estimate these nonlinear effects could be tricky.